Chapter 2

State of the art

This chapter discusses the state of the art in the areas of WebAssembly, Diversification and Runtime Randomization. We present a summary of the relevant related work and the key concepts and background that we use along this writing. We select the discussed works by their novelty and key insights aimed to provide automatic diversification.

TODO Chapter layout

2.1 WebAssembly

In this section we introduce an overview on the motivation for WebAssembly and its usage. Besides, we describe the process to obtain Wasm programs and how this novel technology evolves from being only-browser based to standalone executions in the backend. Nevertheless, we describe its major limitations regarding security that are our main motivation for our research.

The WebAssembly language was first publicly announced in 2015. WebAssembly is a binary instruction format for a stack-based virtual machine. It is designed to address the problem of safe, fast, portable and compact low-level code on the Web. A paper by Haas et al. [24] formalizes the language and its type system. Since 2015, major web browsers have implemented support for the standard.

WebAssembly binaries are usually compiled from source code like C/C++ or Rust [6]. The WebAssembly code is further interpreted or compiled ahead of time into machine code by engines such as the browsers. Since version 8, LLVM supports Wasm as a backend opening the door for its wide collection of frontend languages. The LLVM support was encouraged by the seminal work of Zakai et al. with Emscripten. Emscripten is an open source tool for compiling C/C++ to the WebAssembly. It uses LLVM to create Wasm, but it provides support for faster linking to the object files. Instead of all the IR being compiled by LLVM, the object file is prelinked with Wasm, which is faster. The last version of Emscripten also uses the Wasm LLVM backend as the target for the input code.

WebAssembly for backend execution

Further browser context, the adoption of WebAssembly for backend has grown exponentially in the last four years. For instance, platforms such as Cloudflare and Fastly adapted their platforms to provide FaaS directly with WebAssembly. In 2019, the bytecode alliance team ¹ proposed WebAssembly System Interface (WASI). WASI is the foundation to build Wasm code outside of browser with a system interface platform. WASI allows the adoption of WebAssembly outside web browsers [15] in heterogeneous platforms like the Edge [4, 13]. Previous studies resulted on huge performance increasing in terms of bandwidth saving, execution and process-on-demand spawning [2, 10]. The words of Solomon Hykes ², the former CEO of docker, show the impact of WASI:

If WASM+WASI existed in 2008, we wouldn't have needed to created Docker. That's how important it is. Webassembly on the server is the future of computing. A standardized system interface was the missing link. Let's hope WASI is up to the task!

WebAssembly security and our motivation for diversification

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WebAssembly is characterized by a robust security model [14]. It should run inside a sandboxed execution environment that provides protection against common security issues such as data corruption, code injection and return oriented programming (ROP). However, WebAssembly is vulnerable under certain conditions.

WebAssembly is not exempt of vulnerabilities either at the execution engine's level [19] or the binary itself [12]. Implementations in both, browsers and standalone runtimes [4], have been found to be vulnerable, opening the door to different attacks. This means that if one environment is vulnerable, all the others are vulnerable in the exact same manner as the same WebAssembly binary is replicated. This is clearly a monoculture problem.

On the other hand, the WebAssembly environment lacks of natural diversity [33]. Compared to the work of Harrand et al. [?], in WebAssembly, we cannot use preexisting and different programs to provide diversification solving monoculture. In fact, according to the work of Hilbig et al. [6], the artificial variants created with one of our works contributes to the half of executable and available WebAssembly binaries in the wild.

The current limitations on security and the lack of preexisting diversity motivate our work on software diversification as one possible mitigation among the wide range of security counter-measures.

¹https://bytecodealliance.org/

²https://twitter.com/solomonstre/status/1111004913222324225

2.2 Diversification and Superoptimization for Superdiversification

Program diversification approaches can be applied at different stages of the development pipeline. In this section we analyze the related works for both static and dynamic diversification. Besides, we motivate the usage of the superoptimization strategies to provide a "superdiversifier" that uses intermediate solutions of the searching of optimal programs to provide program variant.

Static diversification consists in synthesizing, building and distributing different, functionally equivalent, binaries to end users. This aims at increasing the complexity and applicability of an attack against a large population of users [55]. Dealing with code-reuse attacks, Homescu et al. [35] propose inserting NOP instruction directly in LLVM IR to generate a variant with different code layout at each compilation. In this area, Coppens et al. [36] use compiler transformations to iteratively diversify software. The aim of their work is to prevent reverse engineering of security patches for attackers targeting vulnerable programs. Their approach, continuously applies a random selection of predefined transformations using a binary diffing tool as feedback [?] . A downside of their method is that attackers are, in theory, able to identify the type of transformations applied and find a way to ignore or reverse them.

Previous works have attempted to generate diversified variants that are alternated during execution. It has been shown to drastically increase the number of execution traces that a side-channel attack requires succeeding. Amarilli et al. [41] are the first to propose generation of code variants against side-channel attacks **TODO** And? . Agosta et al. [34] and Crane et al. [32] modify the LLVM toolchain to compile multiple functionally equivalent variants to randomize the control flow of software **TODO** How **TODO** Why, while Couroussé et al. [27] implement an assembly-like DSL to generate equivalent code at runtime in order to increase protection against side-channel attacks.

Jackson et al. [39] have explored how to use NOP operations inserted during compiling time to statically diversify programs. **TODO** Doe et al. [?] proposed to use the optimization flags of several compilers to generate semantically equivalent binaries out of the same source code. This, techniques place the compiler in the core of the diversification technique. However, this approach is limited by the number of available flags in the compiler implementation and by the fact that the optimization is applied in all possible places in the code.

TODO Blend to superoptimize

Superoptimization

The search of optimal algorithms to compute a function is older as the first compiler. This problem is commonly solved by using human-written heuristics inside the compiler implementations. However, this solution has limitations. First, the

optimizations are applied to small pieces of code and do not take into account more complex processes like instruction selections, register allocation and target-dependent optimizations. Second, the well-known phase ordering problem [38]. To solve this problem, Massalin et al. [56] proposed a superoptimizer, a statistical method to exhaustively explore all possible program constructions to find the smallest program. Given an input program, code superoptimization focuses on searching for a new program variant which is faster or smaller than the original code, while preserving its correctness [28]. The search space for the optimal program is defined by choosing a subset of the machine's instruction set and generating combinations of optimized programs, sorted by length in ascending order. If any of these programs are found to perform the same function as the source program, the search halts. However, for larger instruction sets, the exhaustive exploration approach becomes virtually impossible. Because of this, the paper proposes a pruning method over the search space and a fast probabilistic test to check programs equivalence.

Appart from recent works on the area of Machine Learning [3], to the best of our knowledge, there are two main implementations for superoptimizers using two completely different strategies. Churchill et al. [25] implement STOKE [49] to superoptimize large programs for the Google Native Client stack. They use a bounded verifier to make sure that every generated optimization goes through all the checks for semantic equivalence. STOKE uses a probabilistic approach, following a MonteCarlo-Markov-Chain strategy to select code transformations that lead to smaller programs. On the other hand, Souper [49] automatically generates smaller programs for LLVM following an exhaustive enumerative synthesizis. Souper finds subexpressions at the LLVM function level, builds all possible expression that can be constructed from all the instructions on its own intermediate representation that are no larger than the original subexpression. When Souper finds a replacement, it uses an SMT solver [45] to verify the semantic equivalence with the original program. Superoptimization is time expensive compared to traditional optimization heuristics in compilers, yet, provides deeper and more robust code transformations.

Superdiversification and statement of novelty

During the finding of optimized code, the idea and the implementations of superoptimization discard intermediate solutions that are semantically equivalent to the original program. The discarding of intermediate solutions follows the principle of optimization, finding the best possible program. Jacob et al. [44] propose the use of a "superdiversification" technique, inspired by superoptimization, to synthesize individualized versions of programs, their main idea is to keep the intermediate solutions finding the optimal program. The tool developed by Jacob et al. does not output only the optimal instruction sequence, but any semantically equivalent sequences. Their work focuses on a specific subset of x86 instructions.

In this research we contribute to the state of the art of artificially creating diversity. While the number of related work for software diversity is enormous, none approach has been applied to the context of WebAssembly. One of our contributions, CROW, extrapolates the idea of superdiversification for WebAssembly. CROW works directly with LLVM IR, enabling it to generalize to more languages and CPU architectures something not possible with the x86-specific approach of previous works. Furthermore, we conducted the first sanity check for diversification preservation, researching to what extent browser compilers do not remove introduced diversity.

CROW also can be used in fuzzing campaigns [?] to provide reliability. The diversification created by CROW can unleash hidden behaviors in compilers and interpreters. By generating several functionally equivalent, and yet different variants, deeper bugs can be discovered. Thanks to CROW, a bug was discovered in the Lucet compiler ³. Fastly acknowledged our work as part of a technical blog post ⁴ that describes the bug and the patch.

2.3 Runtime diversification

In this section we describe and discuss the foundation that supports the composition of diverse, yet semantically equivalent, programs to enforce security.

A randomization technique creates a set of unique executions for the very same program [52]. Seminal works include instruction-set randomization [51, 53] to create a unique mapping between artificial CPU instructions and real ones. This makes it very hard for an attacker ignoring the key to inject executable code. This breaks the predictability of program execution and to this extent mitigates certain exploits.

Chew and Song [54] target operating system randomization. They randomize the interface between the operating system and the user applications: the system call numbers, the library entry points (memory addresses) and the stack placement. All those techniques are dynamic, done at runtime using load-time preprocessing and rewriting. Bathkar et al. [52, 50] have proposed three kinds of randomization transformations: randomizing the base addresses of applications and libraries memory regions, random permutation of the order of variables and routines, and the random introduction of random gaps between objects. Dynamic randomization can address different kinds of problems. In particular, it mitigates a large range of memory error exploits. Recent work in this field include stack layout randomization against data-oriented programming [18] and memory safety violations [5], as well as a technique to reduce the exposure time of persistent memory objects to increase the frequency of address randomization [9].

³REPO

⁴https://www.fastly.com/blog/defense-in-depth-stopping-a-wasm-compiler-bug-before-it-became-a-problem

Moving Target Defense and Multivariant execution

Moving Target Defense (MTD) for software was first proposed as a collection of techniques that aim to improve security of a system by constantly moving its vulnerable components [8]. Usually, MTD techniques revolve around changing system inputs and configurations to reduce attack surfaces. This increases uncertainty for attackers and makes their attacks more difficult. Ultimately, potential attackers cannot hit what they cannot see. MTD can be implemented in different ways, including via dynamic runtime platforms [16]. Segupta et al. illustrated how a dynamic MTD system [23] can be applied to different technology stacks. Using this technique, the authors illustrated that some CVE related to specific database engines can be avoided.

On the same topic, Multivariant Execution (MVE) can be seen as a Moving Target Defense strategy. In 2006, security researchers of University of Virginia have laid the foundations of a novel approach to security that consists in executing multiple variants of the same program. They called this "N-variant systems" [48]. Bruschi et al. [47] and Salamat et al. [46] pioneered the idea of executing the variants in parallel. Subsequent techniques focus on Multivariant Execution (MVE) for mitigating memory vulnerabilities [20] and other specific security problems incl. return-oriented programming attacks [29] and code injection [40]. A key design decision of MVE is whether it is achieved in kernel space [17], in user-space [42], with exploiting hardware features [26], or even through code polymorphism [22]. Finally, one can neatly exploit the limit case of executing only two variants [37? , 30]. Notably, Davi et al. proposed Isomeron [31], an approach for execution-path randomization. Isomeron simultaneously loads the original program and a variant. While the program is running, Isomeron continuously flips a coin to decide which copy of the program should be executed next at the level of function calls. With this strategy a potential attacker cannot predict whether the original or the variant of a program will execute.

Statement of novelty

Researching on MVE in a distributed setting like the Edge [?] has been less researched. Voulimeneas et al. proposed a multivariant execution system by parallelizing the execution of the variants in different machines [1] for sake of efficiency. Since, CROW offers both static and runtime diversity for WebAssembly, we propose an original kind of MVE in the context of edge computing, MEWE. We generate multiple program variants, which we execute on edge computing nodes. We use the natural redundancy of Edge-Cloud computing architectures to deploy an internet-based MVE.

With MEWE, We contribute to the field of randomization, at two stages. First, we automatically generate variants of a given program with CROW, which have different WebAssembly code and still behave the same behavior. Second, we randomly select which variant is executed at runtime, creating a multivariant

execution scheme that randomizes the observable behaviors at each run of the program.

2.4 Conclusions

The applications of Software Diversification have been widely researched, not being the case in the WebAssembly context. With this dissertation we aim to settle the foundation to automatically create diversification.

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